

#### Predictive Adaptive Optics Control for High-Contrast Identification of Space Vehicles and Debris



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#### **Outline**



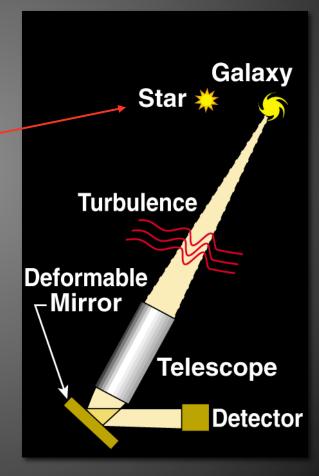
- 1. What is Adaptive Optics?
- 2. A Tomographic Approach to the Fast-Slewing Problem
- 3. Simulated Performance



#### **Laser Guide Star Adaptive Optics**



Use a laser beam to create artificial "star" at altitude of 100 km in atmosphere

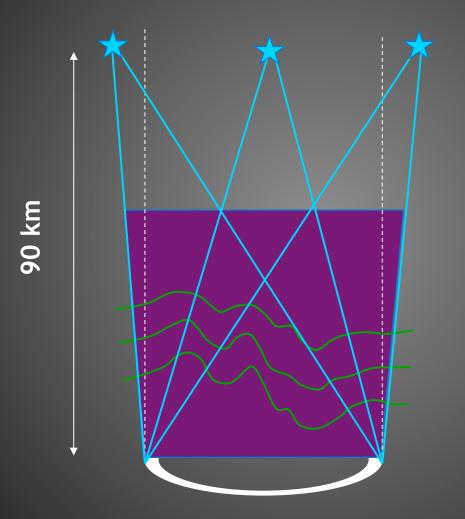


Credit: C. Max



### What is Adaptive Optics Tomography?





Tomography lets you reconstruct turbulence in the entire cylinder of air above the telescope mirror

Tomography is limited by the *finite number* of sampling angles, or limited size of guide star constellation.

Credit: Rigaut, MCAO for Dummies



# Fast-Tracking Scenarios: Active Orbit Modification, Passive Satellite Identification



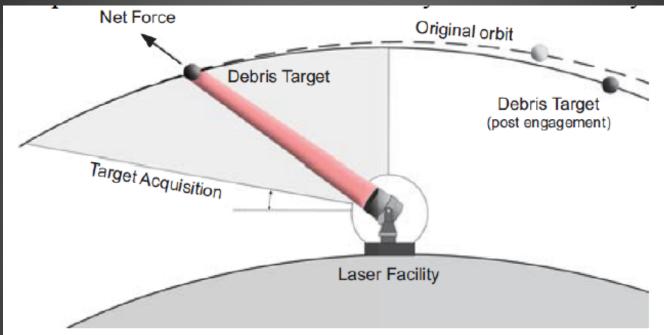


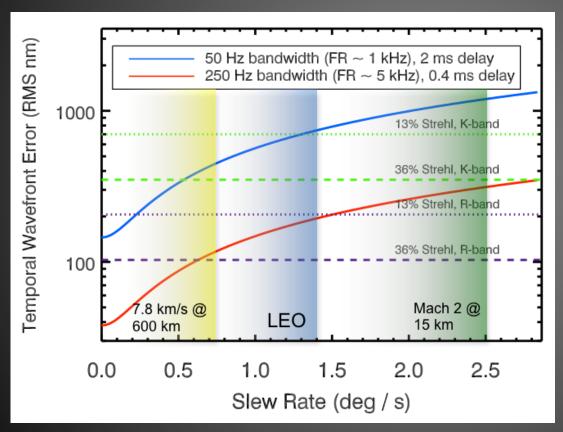
Fig. I: Schematic of laser system for orbital debris collision avoidance.

LightForce: Active Collision Avoidance using photon pressure induced by ground-based lasers. AO is needed to pre-correct laser beams for atmospheric turbulence (Stupl et al. 2011)



## Temporal Errors are Large and Problematic in Fast-Slewing Telescopes



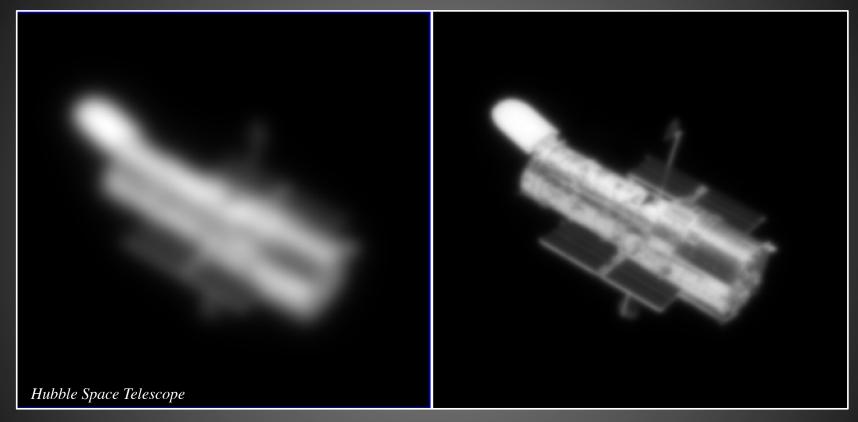


- In fast-tracking AO systems, temporal errors alone preclude diffraction-limited operation at optical/infrared wavelengths for tracking objects in LEO.
- Predictive schemes can reduce temporal errors, making high-strehl and high contrast imaging possible at fast slewing speeds.



## Removal of Temporal Errors Necessary for Reliable Identification





With temporal errors

Removing temporal errors

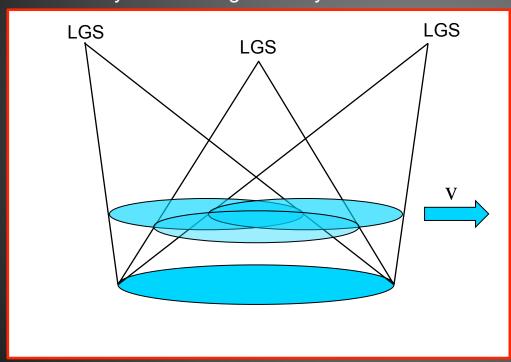
SOR-like 3.5-meter telescope assumed; S = 19% in i-band with prediction; Temporal wavefront errors = 350 nm RMS without prediction; HST-like orbit



### A Tomographic Approach to the Fast-Slewing Problem



#### Geometry for Tracking Two Layers



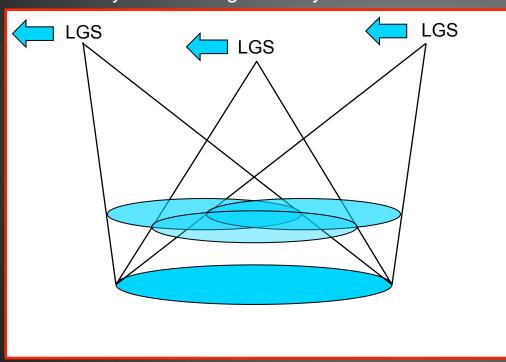
 Example: Two atmospheric layers with static ground layer and upper layer with some velocity



### A Tomographic Approach to the Fast-Slewing Problem



#### Geometry for Tracking Two Layers

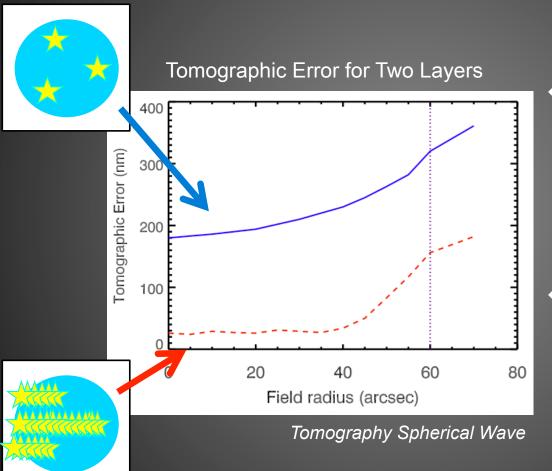


- Geometrically, this is equivalent to an LGS slewing scenario in which the guide stars are tracking across the sky
- The atmospheric phase can be tomographically estimated with essentially zero error



### A Tomographic Approach to the Fast-Slewing Problem





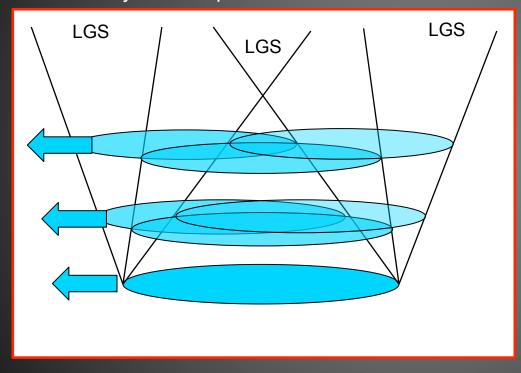
- ◆ For any scenario in which the velocity vector is proportional to layer height, the ensemble of previous iterations can be analyzed tomographically, approximating a dense sampling of guide stars.
- This should be generalizable to cases in which layer height is not proportional to velocity.



### Simulation Design



#### Three-Layer Atmosphere



We simulate an 3.5-m telescope with tomography:

- 3 LGSs over 120" diameter
- 3-layer Taylor frozen-flow atmospheric model assumed at 0, 5, 10 km (55%, 30%, 15%)
- Wind velocities randomized, 0-15 m/s
- 200 realizations of 1 second length, 1 kHz operation
- To isolate the effect of prediction on tomographic error, we add no WFS noise



## **Back-Projection Tomography**



Minimum Variance Back-Projection Tomography (Gavel 2004)

$$\mathbf{v}_{k+1} = \mathbf{v}_k + \Delta \mathbf{v}_k$$

$$\Delta \mathbf{v}_k = a\mathbf{C}\mathbf{e}_k$$

$$\mathbf{e}_k = \mathbf{y} - (\mathbf{A}\mathbf{P}\mathbf{A}^T + \mathbf{N}) \mathbf{v}_k$$

$$\mathbf{x} = \mathbf{P}\mathbf{A}^T \mathbf{v}_{\infty}$$

$$\mathbf{x} = \mathbf{P}\mathbf{A}^T (\mathbf{A}\mathbf{P}\mathbf{A} + \mathbf{N})^{-1} \mathbf{y}$$

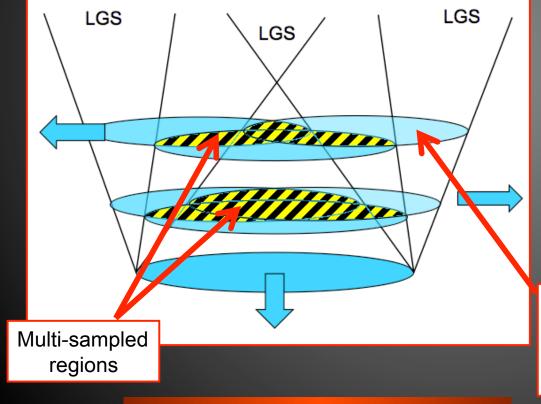
• 25 iterations per time step, alternating preconditioned conjugate gradient / linear steps



## Correction Scheme – Shift & Average Multi-sampled Voxels



$$\Phi'(\mathbf{r}, t') = \frac{1}{n' - n_0 + 1} \sum_{n=n_0}^{n'} \Phi(\mathbf{r} - c(n' - n)\mathbf{v}, cn)$$



- For each layer, replace voxels in downwind direction with shifted and averaged voxels from tomographic time history
- Only shift voxels originating in multi-sampled region, where height can be effectively determined
- Wind vectors assumed to be known perfectly

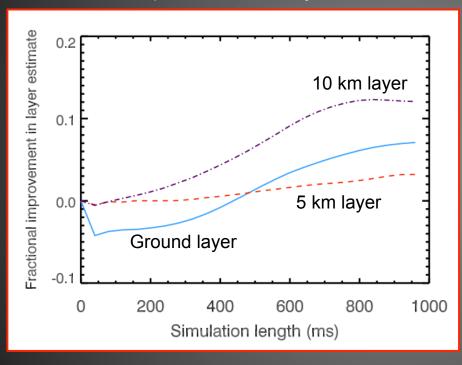
Phase height cannot be constrained in sparsely-sampled regions from tomography alone!



## Prediction Improves RMS Errors on Layer Estimates



#### Fractional Improvement in Layer Estimates



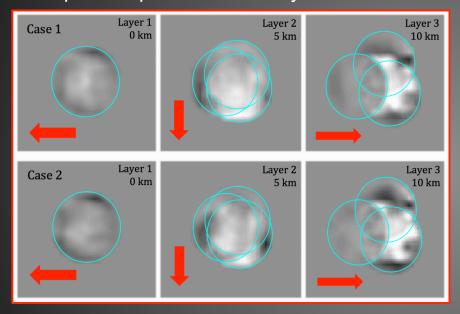
- After 1 second, on average, the layer estimates improve 3-13%
- Downwind regions improve 10-30%, especially for high altitude layers



## Prediction Improves RMS Errors on Layer Estimates



#### Maps of Improvement in Layer Estimates

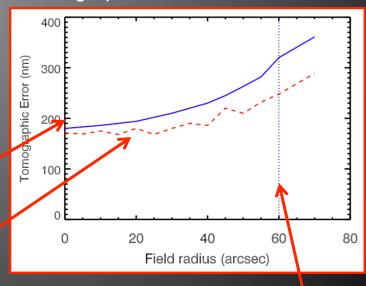


Without shift & average

With shift & average

 With downwind layers better determined, the tomographic error improves beyond the radius of the guide stars

#### Tomographic Error vs. Field Radius



LGS radius



### **Conclusions from Simulations**



- Fast-slewing telescopes plagued by temporal errors when tracking LEO objects
- A tomographic predictive approach could improve wavefront errors and contrast of fast-slewing telescopes
- 3. In simulation, shifting-and-averaging predictive control provides 3-13% benefits in tomographic wavefront estimation quality at all atmospheric layers (temporal errors not tested)